

Three-Level (and Crossed) Longitudinal Models

- Topics:
 - **Decomposing variation across three levels in clustered longitudinal data**
 - Unconditional (time only) model specification
 - Conditional (other predictors) model specification
 - Other kinds of three-level and crossed designs

What determines the number of levels?

- **Answer: the model for the outcome variance ONLY**
- How many dimensions of sampling in the outcome?
 - Time within person → 2-level model
 - Time within person within family → 3-level model
 - Time within person within family within country → 4-level model
 - Sampling dimensions may also be crossed instead of nested, or the top dimension may be modeled with fixed effects instead
- Need at least one pile of variance per dimension (for 3 levels, that's 2 sets of random effects and a residual)
 - Include whatever predictors you want for each level, but keep in mind that the usefulness of your predictors will be constrained by how much Y variance exists in its relevant sampling dimension

Empty Means, 3-Level Random Intercept Model

Notation: t = level-1 time, i = level-2 person, j = level-3 group

Level 1: $y_{tij} = \beta_{0ij} + e_{tij}$

Residual = time-specific deviation
from person's predicted outcome

Level 2: $\beta_{0ij} = \delta_{00j} + U_{0ij}$

Person Random Intercept
= person-specific deviation
from group's predicted outcome

Level 3: $\delta_{00j} = Y_{000} + V_{00j}$

Fixed Intercept
= grand mean
(because no
predictors yet)

Group Random Intercept
= group-specific deviation
from fixed intercept

3 Total Parameters:

Model for the Means (1):

- Fixed Intercept Y_{000}

Model for the Variance (2):

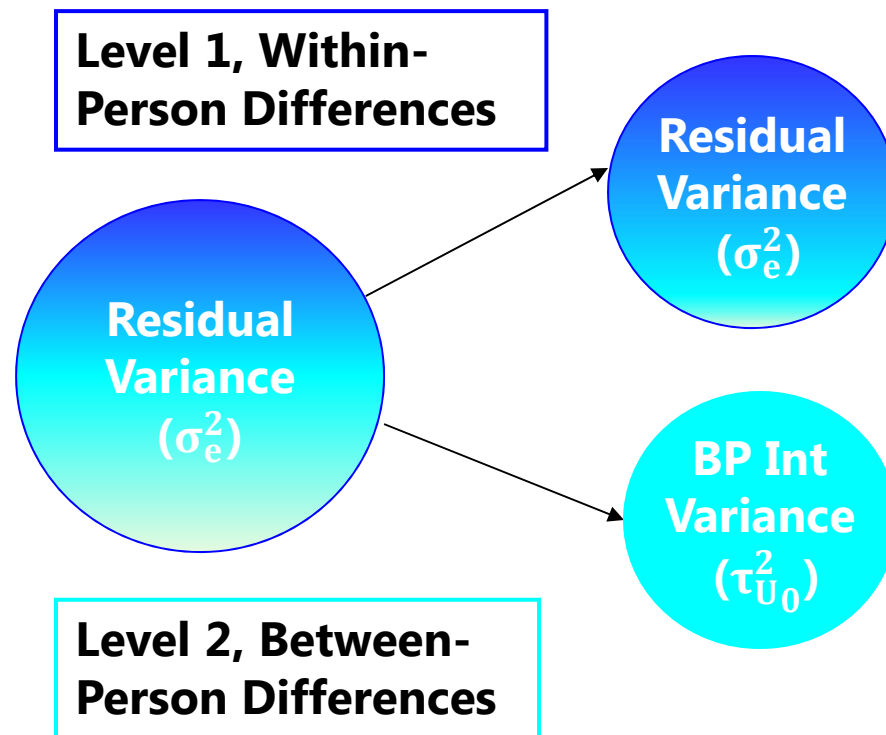
- Level-1 Variance of $e_{tij} \rightarrow \sigma_e^2$
- Level-2 Variance of $U_{0ij} \rightarrow \tau_{U_0}^2$
- Level-3 Variance of $V_{00j} \rightarrow \tau_{V_{00}}^2$

Composite equation:

$$y_{tij} = Y_{000} + V_{00j} + U_{0ij} + e_{tij}$$

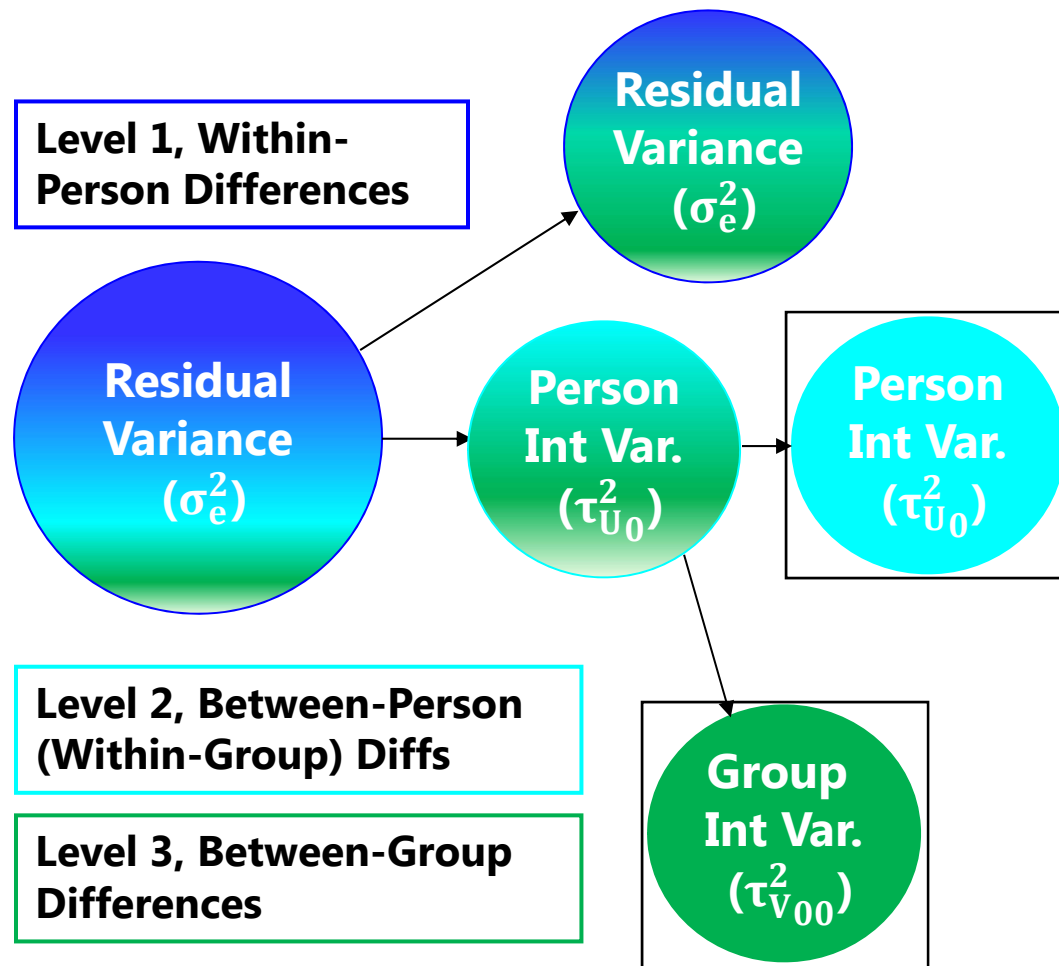
2-Level Random Intercept Model

- Where does each kind of person dependency go? Into a new random effects variance component (or “pile” of variance):
- Let's start with an empty means, random intercept 2-level model for time within person:



3-Level Random Intercept Model

- Now let's see what happens in an empty means, random intercept 3-level model of time within person within groups:



ICCs in a 3-Level Random Intercept Model

Example: Time within Person within Group

- ICC for level 2 (and level 3) relative to level 1:

- $$ICC_{L2} = \frac{\text{Between-Person}}{\text{Total}} = \frac{L3+L2}{L3+L2+L1} = \frac{\tau_{V00}^2 + \tau_{U0}^2}{\tau_{V00}^2 + \tau_{U0}^2 + \sigma_e^2}$$

→ This ICC expresses similarity of occasions from same person (and by definition, from the same group) → of the **total variation in Y**, how much of it is **between persons, or not due to time?**

- ICC for level 3 relative to level 2 (ignoring level 1):

- $$ICC_{L3} = \frac{\text{Between-Group}}{\text{Between-Person}} = \frac{L3}{L3+L2} = \frac{\tau_{V00}^2}{\tau_{V00}^2 + \tau_{U0}^2}$$

See also what I call:
 $ICC_{L3B} = L3/\text{total}$
(as given by STATA)

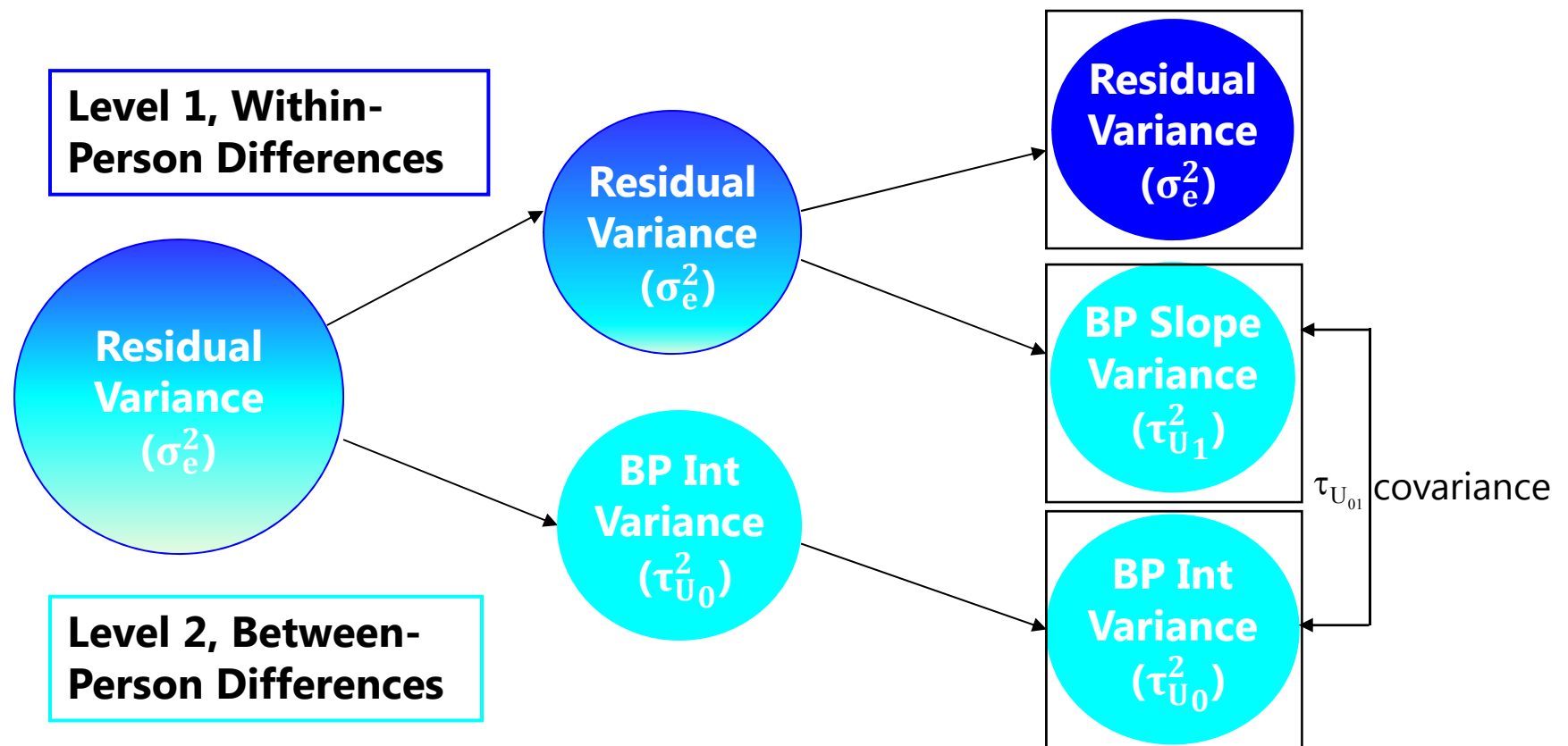
→ This ICC expresses similarity of persons from same group (ignoring within-person variation over time) → of **that total between-person variation in Y**, how much of that is actually **between groups?**

Three-Level (and Crossed) Longitudinal Models

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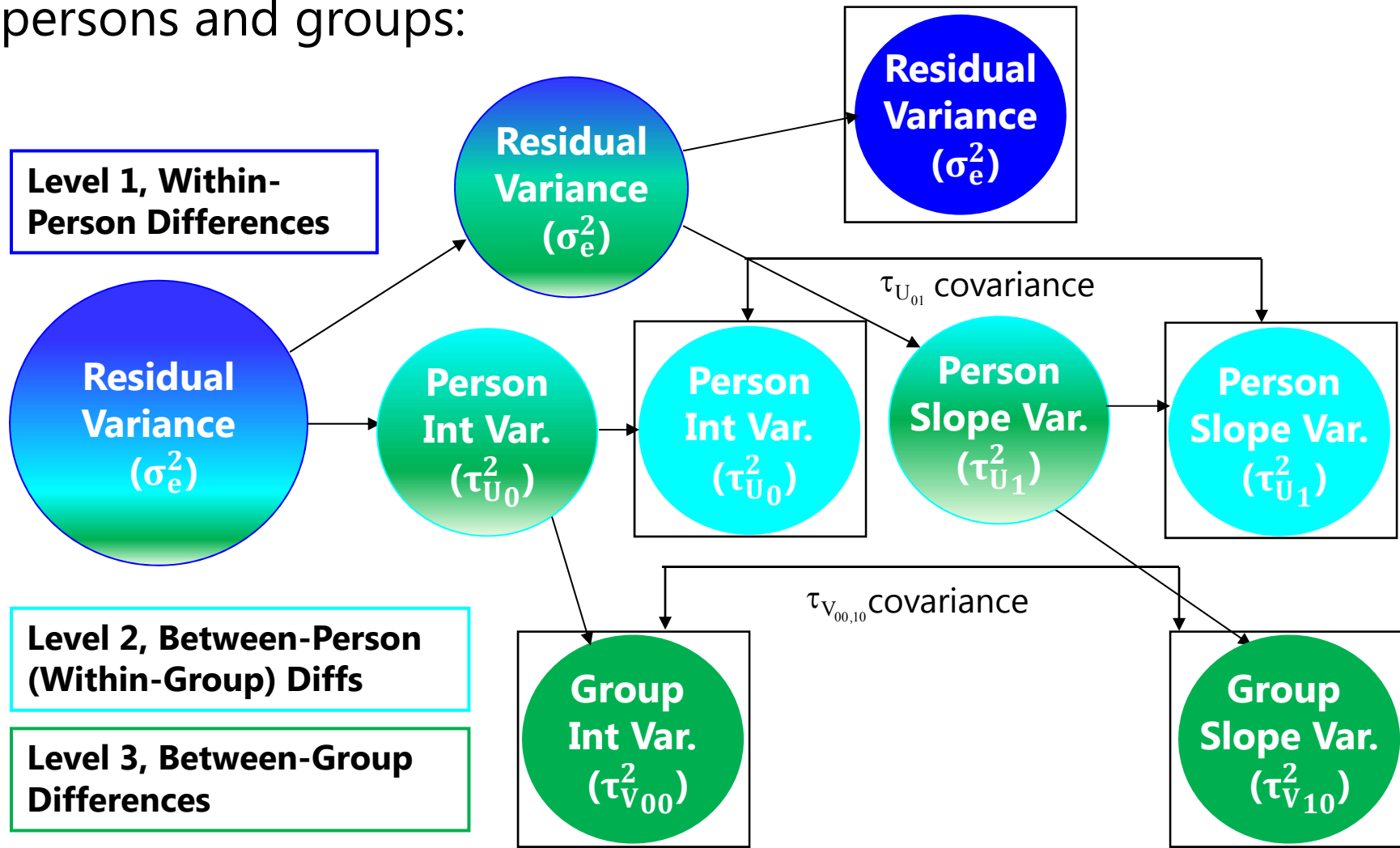
2-Level Random Slope Model

- What about time? After adding fixed effects of time, we can add random effects of time over persons in a 2-level model:

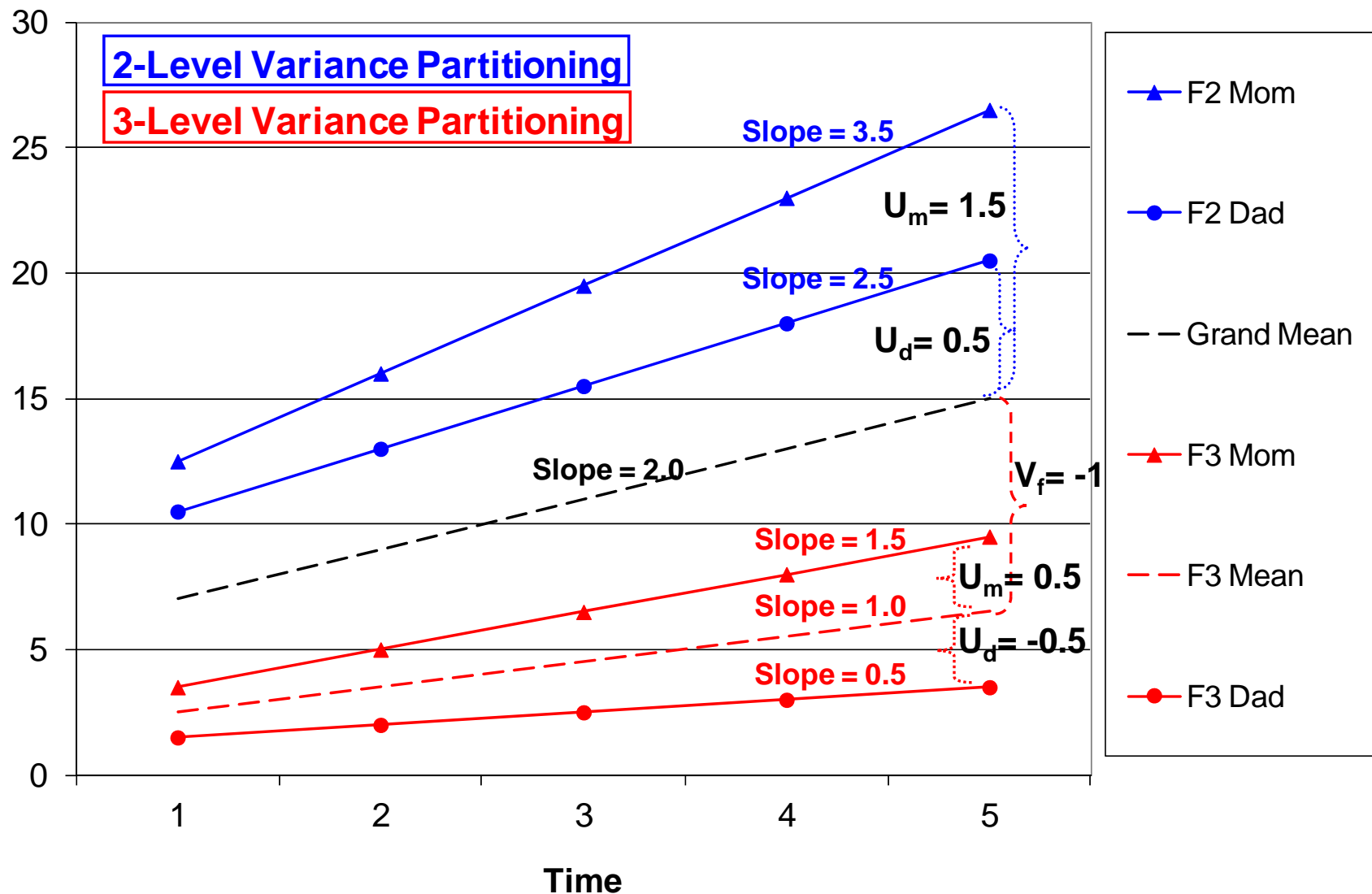


3-Level Random Slope Model

- In a 3-level model, we can have random effects of time over persons and groups:



Random Time Slopes at both Levels 2 AND 3? An example with family as group:



3-Level Random Time Slope Model

Notation: t = level-1 time, i = level-2 person, j = level-3 group

Level 1: $y_{tij} = \beta_{0ij} + \beta_{1ij}(\text{Time}_{tij}) + e_{tij}$ ← Residual = time-specific deviation from person's predicted growth line (σ_e^2)

Level 2: $\beta_{0ij} = \delta_{00j} + U_{0ij}$
 $\beta_{1ij} = \delta_{10j} + U_{1ij}$ ← Person Random Intercept and Slope = person-specific deviations from group's predicted intercept, slope ($\tau_{U0}^2, \tau_{U1}^2, \tau_{U01}$)

Level 3: $\delta_{00j} = Y_{000} + V_{00j}$
 $\delta_{10j} = Y_{100} + V_{10j}$ ← Group Random Intercept and Slope = group-specific deviations from fixed intercept, slope ($\tau_{V00}^2, \tau_{V10}^2, \tau_{V00,10}$)

**Fixed Intercept,
Fixed Linear
Time Slope**

Composite equation (9 parameters):

$$y_{tij} = (Y_{000} + V_{00j} + U_{0ij}) + (Y_{100} + V_{10j} + U_{1ij})(\text{Time}_{tij}) + e_{tij}$$

ICCs for Random Intercepts and Slopes

- Once random slopes are included at both level-3 and level-2, ICCs can be computed for the random intercepts and slopes specifically (which would be the level-3 type of ICC)

$$ICC_{Int} = \frac{\text{Between} - \text{Group}}{\text{Between} - \text{Person}} = \frac{\text{L3 Int}}{\text{L3 Int} + \text{L2 Int}} = \frac{\tau_{V00}^2}{\tau_{V00}^2 + \tau_{U0}^2}$$

$$ICC_{Slope} = \frac{\text{Between} - \text{Group}}{\text{Between} - \text{Person}} = \frac{\text{L3 Slope}}{\text{L3 Slope} + \text{L2 Slope}} = \frac{\tau_{V10}^2}{\tau_{V10}^2 + \tau_{U1}^2}$$

- Can be computed for any level-1 slope that is random at both levels (e.g., linear and quadratic time, time-varying predictors)
- Be careful when the model is uneven across levels, though

$$\frac{\text{Random Level 2: int, linear, quad}}{\text{Random Level 3: int, linear}} \rightarrow \frac{\text{Linear is when time} = 0}{\text{Linear is at any occasion}}$$

More on Random Slopes in 3-Level Models

- Any level-1 predictor can have a random effect over level 2, level 3, or over both levels, but I recommend working your way **UP the higher levels** for assessing random effects...
 - e.g., Does the effect of time vary over persons?
 - If so, does the effect of time vary over groups, too? → Is there a commonality in how people from the same group change over time?
- ... because random effects at level 3 only are possible but unlikely (e.g., means everyone in the group changes the same)
- Level-2 predictors can also have random effects over level 3
 - e.g., Does the effect of a person characteristic vary over groups?
- Level-1, level-2, and level-1 by level-2 cross-level interactions can all have random effects over level 3, too
 - But tread carefully! The more random effects you have, the more likely you are to have convergence problems ("**G** matrix not positive definite")

Three-Level (and Crossed) Longitudinal Models

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Conditional Model Specification

- Remember separating between- and within-person effects?
Now there are 3 potential effects for any level-1 predictor!
 - Example: Effect of stress on wellbeing, both measured over time within person within families:
 - **Level 1** (Time): During **Times** of more stress, people have lower (time-specific) wellbeing than in times of less stress
 - **Level 2** (Person): **People** in the family who have more stress have lower (person average) wellbeing than people in the family who have less stress
 - **Level 3** (Family): **Families** who have more stress have lower (family average) wellbeing than families who have less stress
- 2 potential effects for any level-2 predictor, also
 - Example: Effect of baseline level of person coping skills in same design:
 - **Level 2** (Person): **People** in the family who cope better have better (person average) wellbeing than people in the family who cope worse
 - **Level 3** (Family): **Families** who cope better have better (family average) wellbeing than families who cope worse

Separate Total Effects Per Level Using Variable-Based-Centering

- **Level 1 (Time):** *Time-varying stress relative to person mean*
 - $WP_{stress_{tij}} = Stress_{tij} - PersonMeanStress_{ij}$
 - Directly tests if within-person effect $\neq 0$?
 - **Total** within-person effect of having more stress **than usual** $\neq 0$?
- **Level 2 (Person):** *Person mean stress relative to family*
 - $WF_{stress_{ij}} = PersonMeanStress_{ij} - FamilyMeanStress_j$
 - Directly tests if within-family effect $\neq 0$?
 - **Total** effect of having more stress **than other family members** $\neq 0$?
- **Level 3 (Family):** *Family mean stress relative to all families (from constant)*
 - $BF_{stress_j} = FamilyMeanStress_j - C$
 - Directly tests if between-family effect $\neq 0$?
 - **Total** effect of having more stress **than other families** $\neq 0$?

Separate Total Effects Per Level Using Variable-Based-Centering

Notation: t = level-1 time, i = level-2 person, j = level-3 group
 PM = person mean, FM = family mean, C = centering constant

Level 1: $y_{tij} = \beta_{0ij} + \beta_{1ij}(\text{Time}_{tij}) + \beta_{2ij}(\text{Stress}_{tij} - \text{PMstress}_{ij}) + e_{tij}$

Level 2: $\beta_{0ij} = \delta_{00j} + \delta_{01j}(\text{PMstress}_{ij} - \text{FMstress}_j) + U_{0ij}$

$$\beta_{1ij} = \delta_{10j} + U_{1ij}$$

$$\beta_{2ij} = \delta_{20j} + (U_{2ij})$$

Level 3: $\delta_{00j} = Y_{000} + Y_{001}(\text{FMstress}_j - C) + V_{00j}$

$$\delta_{01j} = Y_{010} + (V_{01j})$$

$$\delta_{10j} = Y_{100} + V_{10j}$$

$$\delta_{20j} = Y_{200} + (V_{20j})$$

Fixed intercept,
Between-family
stress main effect

Within-family stress main effect

Time main effect

Within-person stress main effect

Contextual Effects Per Level Using Constant-Based-Centering

- **Level 1 (Time):** *Time-varying stress (relative to sample constant)*
 - $TVstress_{tij} = Stress_{tij} - C$
 - Directly tests if within-person effect $\neq 0$?
 - **Total** within-person effect of having more stress **than usual** $\neq 0$?
- **Level 2 (Person):** *Person mean stress (relative to sample constant)*
 - $BPstress_{ij} = PersonMeanStress_{ij} - C$
 - Directly tests if within-person and within-family effects $\neq ?$
 - **Contextual** effect of having more stress **than other family members** $\neq 0$?
- **Level 3 (Family):** *Family mean stress relative to all families (from constant)*
 - $BFstress_j = FamilyMeanStress_j - C$
 - Directly tests if within-family and between-family effects $\neq ?$
 - **Contextual** effect of having more stress **than other families** $\neq 0$?

Contextual Effects Per Level Using Constant-Based-Centering

Notation: t = level-1 time, i = level-2 person, j = level-3 group
 PM = person mean, FM = family mean, C = centering constant

Level 1: $y_{tij} = \beta_{0ij} + \beta_{1ij}(\text{Time}_{tij}) + \beta_{2ij}(\text{Stress}_{tij} - C) + e_{tij}$

Level 2: $\beta_{0ij} = \delta_{00j} + \delta_{01j}(\text{PMstress}_{ij} - C) + U_{0ij}$

$$\beta_{1ij} = \delta_{10j} + U_{1ij}$$

$$\beta_{2ij} = \delta_{20j} + (U_{2ij})$$

Level 3: $\delta_{00j} = Y_{000} + Y_{001}(\text{FMstress}_j - C) + V_{00j}$

$$\delta_{01j} = Y_{010} + (V_{01j})$$

$$\delta_{10j} = Y_{100} + V_{10j}$$

$$\delta_{20j} = Y_{200} + (V_{20j})$$

Fixed intercept,
Contextual family
 stress main effect

Contextual within-family stress main effect

Time main effect

Within-person stress main effect

What does it mean to omit higher-level effects under each centering method?

- **Variable-Based-Centering:** Removing terms means the effect at that level does not exist (= 0)
 - Remove L3 effect? Assume L3 Between-Family effect = 0
 - *L1 effect = Within-Person effect, L2 effect = Within-Family effect*
 - Then remove L2 effect? Assume L2 Within-Family effect = 0
 - *L1 effect = Within-Person effect*
- **Constant-Based-Centering:** Removing terms means the effect at that level is equivalent to the effect at the level beneath it
 - Remove L3 effect? Assume L3 Between-Family = L2 Within-Family effect
 - *L1 effect = Within-Person effect, L2 effect = 'smushed' WF and BF effects*
 - Then remove L2 effect? Assume L2 Between-Person effect = L1 effect
 - *L1 'smushed' = Within-Person, Within-Family, and Between-Family effects*

Interactions belong at each level, too...

- Example: Is the effect of stress on wellbeing moderated by time-invariant person coping? Via variable-based-centering...

- **Stress Effects**

- **Level 1 (Time):** $WPstress_{tij} = Stress_{tij} - PersonMeanStress_{ij}$
- **Level 2 (Person):** $WFstress_{ij} = PersonMeanStress_{ij} - FamilyMeanStress_j$
- **Level 3 (Family):** $BFstress_j = FamilyMeanStress_j - C$

- **Coping Effects**

- **Level 2 (Person):** $WFcope_{ij} = Cope_{ij} - FamilyMeanCope_j$
- **Level 3 (Family):** $BFcope_j = FamilyMeanCope_j - C$

- **Interaction Effects**

- With level 1 stress: $WPstress_{tij} * WFcope_{ij}$, $WPstress_{tij} * BFcope_j$
- With level 2 stress: $WFstress_{ij} * WFcope_{ij}$, $(WFstress_{ij} * BFcope_j)$
- With level 3 stress: $BFstress_j * BFcope_j$, $(BFstress_j * WFcope_{ij})$

Interactions belong at each level, too...

Notation: t = level-1 time, i = level-2 person, j = level-3 group
 PM = person mean, FM = family mean, C = centering constant

$$\text{Level 1: } y_{tij} = \beta_{0ij} + \beta_{1ij}(\text{Time}_{tij}) + \beta_{2ij}(\text{Stress}_{tij} - \text{PMstress}_{ij}) + e_{tij}$$

$$\begin{aligned} \text{Level 2: } \beta_{0ij} &= \delta_{00j} + \delta_{01j}(\text{PMstress}_{ij} - \text{FMstress}_j) \\ &\quad + \delta_{02j}(\text{Cope}_{ij} - \text{FMcope}_j) \\ &\quad + \delta_{03j}(\text{PMstress}_{ij} - \text{FMstress}_j)(\text{Cope}_{ij} - \text{FMcope}_j) + U_{0ij} \\ \beta_{1ij} &= \delta_{10j} + U_{1ij} \\ \beta_{2ij} &= \delta_{20j} + \delta_{21j}(\text{Cope}_{ij} - \text{FMcope}_j) + (U_{2ij}) \end{aligned}$$

$$\begin{aligned} \text{Level 3: } \delta_{00j} &= Y_{000} + Y_{001}(\text{FMstress}_j - C) + Y_{002}(\text{FMcope}_j - C) \\ &\quad + Y_{003}(\text{FMstress}_j - C)(\text{FMcope}_j - C) + V_{00j} \\ \delta_{01j} &= Y_{010} + (V_{01j}) & \delta_{02j} &= Y_{020} + (V_{02j}) & \delta_{03j} &= Y_{030} + (V_{03j}) \\ \delta_{10j} &= Y_{100} + V_{10j} \\ \delta_{20j} &= Y_{200} + Y_{202}(\text{FMcope}_j - C) + (V_{20j}) & \delta_{21j} &= Y_{210} + (V_{21j}) \end{aligned}$$

Summary: Clustered Longitudinal Models

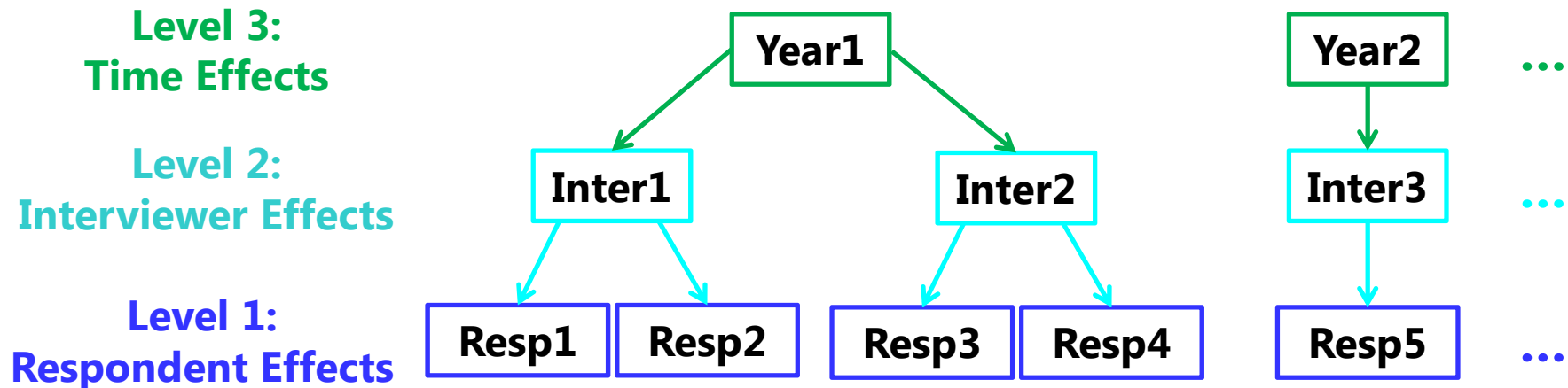
- Estimating 3-level models requires no new concepts, but everything is just at an order of complexity higher:
 - Proportioning variance over 3 levels instead of 2 → 2+ ICCs
 - Random slope variance will come from term directly beneath:
 - Level-2 random slope comes from level-1 residual
 - Level-3 random slope comes from level-2 random slope (or residual)
 - Level-1 effects can be random over level 2, level 3, or both
 - ICCs can be computed for level-1 slopes that are random over both level-2 and level-3 (assuming the L2 and L3 models match)
 - Convergence of level-1 effects should be tested over levels 2 AND 3
 - Level-2 effects can be random over level 3
 - Convergence of level-2 effects should be tested over level 3
 - Level-3 effects cannot be random; no convergence testing needed
 - Phew....

Three-Level (and Crossed) Longitudinal Models

- Topics:
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 - Conditional (other predictors) model specification
 - **Other kinds of three-level and crossed designs**

Other 3-Level Designs

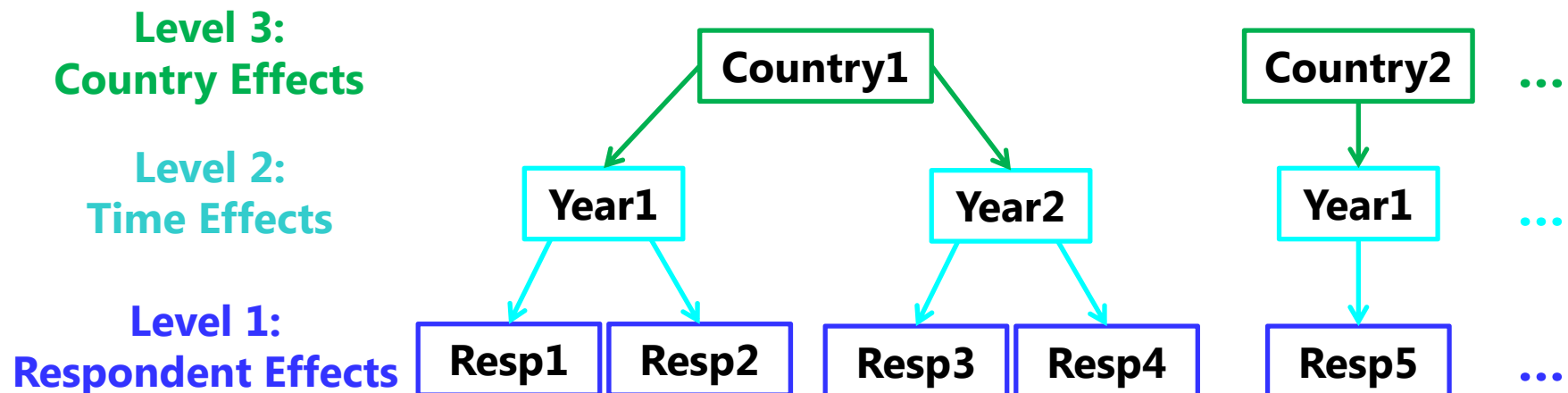
- The sampling design for the outcome (not the predictors) dictates what your levels will be, **so time may not always be level 1**
- Example: Predicting answer compliance in respondents nested in interviewers, collected over several years (all different people)



- Based on the sampling of time, time may be modeled...
 - As fixed effects in the model for the means → 2-level model instead
 - Best to use dummy codes for time if few occasions OR no time-level predictors of interest
 - As a random effect in the model for the variance → 3-level model
 - Then differences in compliance rates over time can be predicted by time-level predictors

Other 3-Level Designs

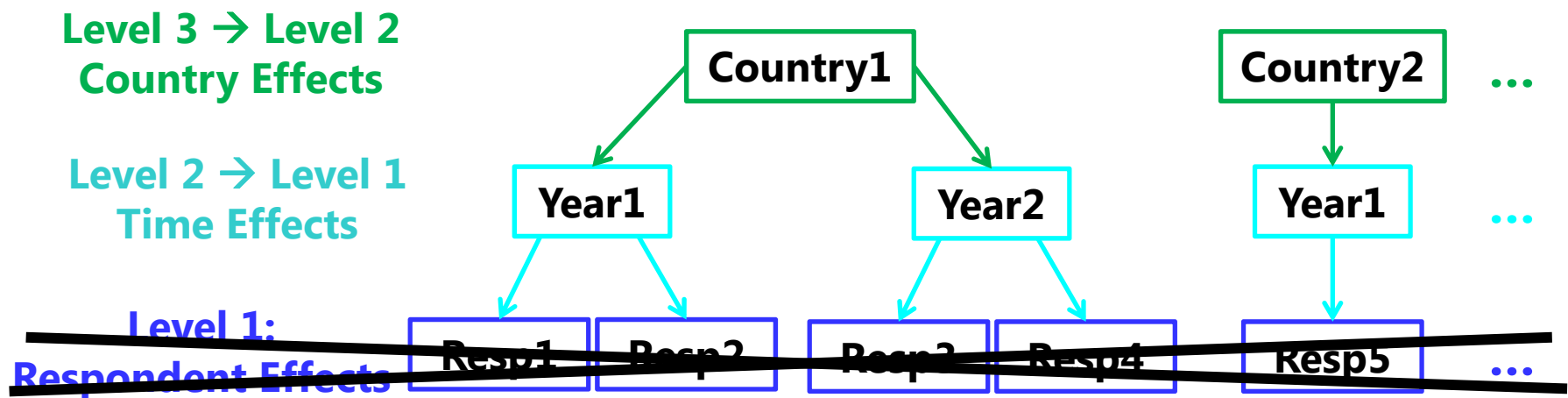
- Another example: Predicting **time-specific respondent outcomes** for people nested in countries, collected over several years (all different people, but the same countries measured over time)



- Before including any fixed effects of time, country and time are actually crossed, not nested as shown here
 - Are nested after controlling for which occasion is which via fixed effects (using dummy codes per mean or a time trend that describes the means)
 - Time is still a level because not all countries change the same way

3-Level Designs: Predictors vs. Outcomes

- Same example: What if, instead of respondent outcomes, we wanted to predict **time-varying country outcomes**?

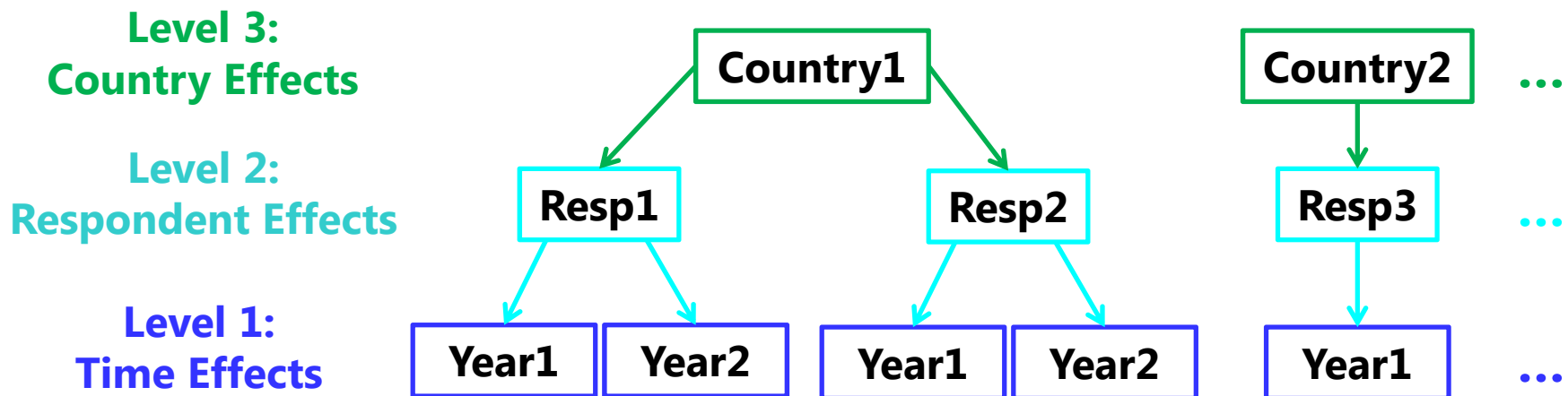


Because the outcome was measured at level 2 (country per time):

- Respondents are no longer a level at all (no outcomes for them)
- So there is nothing for respondent predictors to do, except at higher levels
 - **Time-specific averages** of respondent predictors → time-level outcome variation
 - **Across time, country averages** of respondent predictors → country-level outcome variation

Other 3-Level Designs: Predictors by Level

- Last example: Predicting **time-specific respondent outcomes** for people nested in countries, collected over several years (all same people and same countries are measured over time)



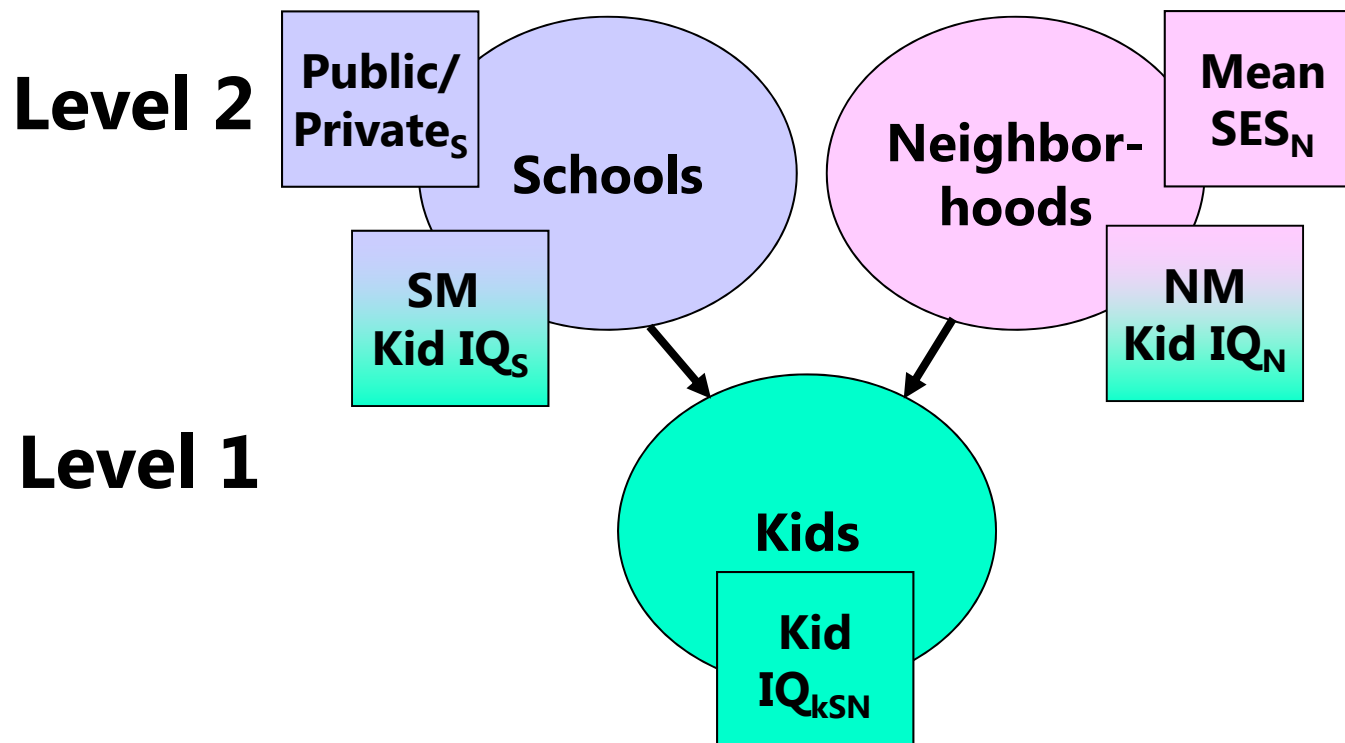
- Country predictors can be included at level 3 only (no random effects)
- Person predictors should be included at levels 2 and 3 (+random over 3)
- What about effects of time-varying predictors?
 - For People: effects should be included at all 3 levels (+random over 2 and 3)
 - For Countries: effects are only possible at levels 1 and 3 (+random over 3)

More Complex Multilevel Designs

- Multilevel models are specified based on the relevant dimensions by which observations differ each other, and how the units are organized
- Two-level models have at least two piles of variance, in which level-1 units are nested within level-2 units:
 - Longitudinal Data: Time nested within Persons
 - Students nested within Classes
- Three-level models have at least three piles of variance, in which level-2 units are nested within level-3 units:
 - Time nested within Persons within Families
 - Student nested within Classes within Schools
- In other designs, multiple sources of systematic variation may be present, but the sampling may be crossed instead...
 - Same idea as crossed random effects (i.e., as for persons and items), but these are known as “cross-classified” models in the clustered data world
 - Here are a few examples on when this might happen...

Kids (k), Schools (s), and Neighborhoods (n)

- Kids are nested within schools AND within neighborhoods
- Not all kids from same neighborhood live in same school, so schools and neighborhoods are crossed at level 2
- Can include predictors for each source of variation



Kids (k), Schools (s), and Neighborhoods (n)

$$\begin{aligned} y_{kSN} = & \mathbf{Y}_{000} && \rightarrow \text{fixed intercept (all } x\text{'s} = 0) \\ & + \mathbf{Y}_{010}(\text{Private}_s) + \mathbf{Y}_{020}(\text{SMIQ}_s) && \rightarrow \text{school effects} \\ & + \mathbf{Y}_{001}(\text{SES}_N) + \mathbf{Y}_{002}(\text{NMIQ}_N) && \rightarrow \text{neighborhood effects} \\ & + \mathbf{Y}_{100}(\text{KidIQ}_{kSN}) && \rightarrow \text{kid effects} \\ & + \mathbf{U}_{0s0} && \rightarrow \text{random effect of school} \\ & + \mathbf{U}_{00N} && \rightarrow \text{random effect of neighborhood} \\ & + \mathbf{e}_{kSN} && \rightarrow \text{residual kid-to-kid variation} \end{aligned}$$

Time (t), Students (s), and Classes (c)

- Students are nested within Classes at each occasion...
- But if students move into different classes across time...
 - Time at level 1 is nested within Student AND within Classes
 - Student is crossed with Class at level 2
- How to model a time-varying random classroom effect?
 - This is the basis of so-called “value-added models”
- (At least) Two options via fixed or random effects:
 - Acute effect: Effect for class operates only when kids are in the class
 - e.g., Class effect \leftarrow teacher bias
 - Once a student is out of the class, class effect is no longer present
 - Transfer effect: Effect for class operates now and in the future...
 - e.g., Class effect \leftarrow differential learning
 - Effect stays with the student in the future (i.e., a “layered” value-added model)

Time (t), Students (s), and Classes (c)

- Custom-built intercepts for time-varying effects of classes
 - An intercept is usually a column of 1's, but ours will be 0's and 1's to serve as switches that turn on/off class effects

Student ID	Class ID	Grade	Year	Per-Year Class ID (-99 = missing)			Intercepts for Acute Effects			Intercepts for Transfer Effects		
				Year 0 Class	Year 1 Class	Year 2 Class	Year 0 Intercept	Year 1 Intercept	Year 2 Intercept	Year 0 Effect	Year 1 Effect	Year 2 Effect
101	1	3	0	1	-99	43	1	0	0	1	0	0
101	-99	4	1	1	-99	43	0	0	0	0	0	0
101	43	5	2	1	-99	43	0	0	1	1	0	1
102	3	3	0	3	21	42	1	0	0	1	0	0
102	21	4	1	3	21	42	0	1	0	1	1	0
102	42	5	2	3	21	42	0	0	1	1	1	1

Time (t), Students (s), and Classes (c)

- Hoffman (2014) Equation 11.3: fixed effects model for class as a categorical time-varying predictor:

- Allows for control of classes only....

$$\begin{aligned}\text{Effort}_{tsc} = & \gamma_{000} + \gamma_{100}(\text{Year01}_{tsc}) + \gamma_{200}(\text{Year12}_{tsc}) + U_{0s0} + e_{tsc} \\ & + \gamma_{001}^0(\text{Class1}_c)(\text{Int0}_{tsc}) + \gamma_{002}^0(\text{Class2}_c)(\text{Int0}_{tsc}) \cdots + \gamma_{00C}^0(\text{ClassC}_c)(\text{Int0}_{tsc}) \\ & + \gamma_{001}^1(\text{Class1}_c)(\text{Int1}_{tsc}) + \gamma_{002}^1(\text{Class2}_c)(\text{Int1}_{tsc}) \cdots + \gamma_{00C}^1(\text{ClassC}_c)(\text{Int1}_{tsc}) \\ & + \gamma_{001}^2(\text{Class1}_c)(\text{Int2}_{tsc}) + \gamma_{002}^2(\text{Class2}_c)(\text{Int2}_{tsc}) \cdots + \gamma_{00C}^2(\text{ClassC}_c)(\text{Int2}_{tsc})\end{aligned}$$

- Hoffman (2014) Equation 11.4: class as a random effects crossed with students at level 2:

- Controls and models class-related variance so it can be predicted

$$\begin{aligned}\text{Effort}_{tsc} = & \gamma_{000} + \gamma_{100}(\text{Year01}_{tsc}) + \gamma_{200}(\text{Year12}_{tsc}) + U_{0s0} + e_{tsc} \\ & + U_{00c}^0(\text{Int0}_{tsc}) + U_{00c}^1(\text{Int1}_{tsc}) + U_{00c}^2(\text{Int2}_{tsc})\end{aligned}$$

More on Cross-Classified Models

- In crossed models, lower-level predictors can have random slopes of over higher levels AND random slopes of the other crossed factor at the same level
 - Example: Kids, Schools, and Neighborhoods (data permitting)
 - Kid effects could vary over schools AND/OR neighborhoods
 - School effects could vary over neighborhoods (both level 2)
 - Neighborhood effects could vary over schools (both level 2)
- Concerns about smushing still apply over both level-2's
 - Separate contextual effects of kid predictors for schools and neighborhoods (e.g., after controlling for how smart you are, it matters incrementally whether you go to a smart school AND if you live in a neighborhood with smart kids)

Summary: Nested or Crossed Designs

- Dimensions of sampling can result in systematic differences (i.e., dependency) that needs to be accounted for in the model for the variances
 - Sometimes this dependency is from nested sampling
 - Sometimes this dependency is from crossed sampling
- Multilevel models that include crossed random effects (or cross-classified models):
 - Can address this dependency (statistical motivation)
 - Can quantify and predict the amount of variation due to each source (substantive motivation)
 - Can include simultaneous hypothesis tests pertaining to each source of variation (substantive motivation)